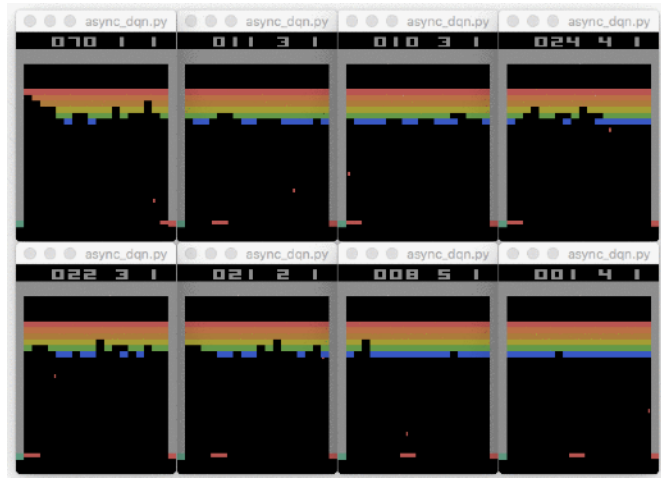


# Introduction to Artificial Intelligence

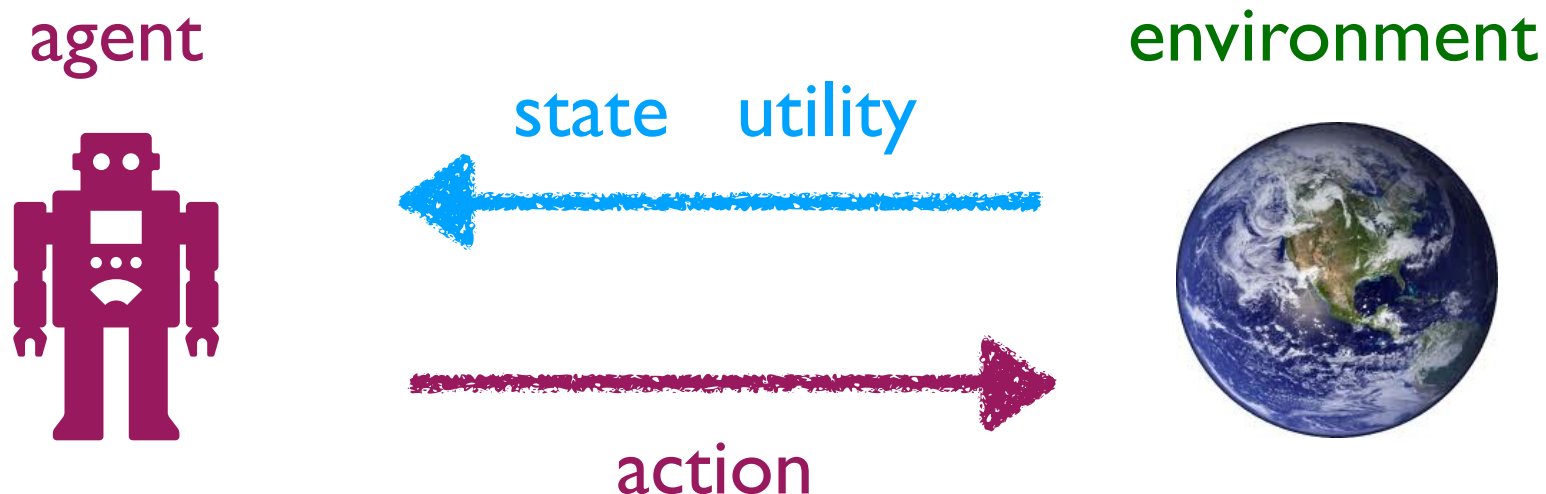
丁尧相  
浙江大学

Fall & Winter 2022  
Week 9

# Decision Making

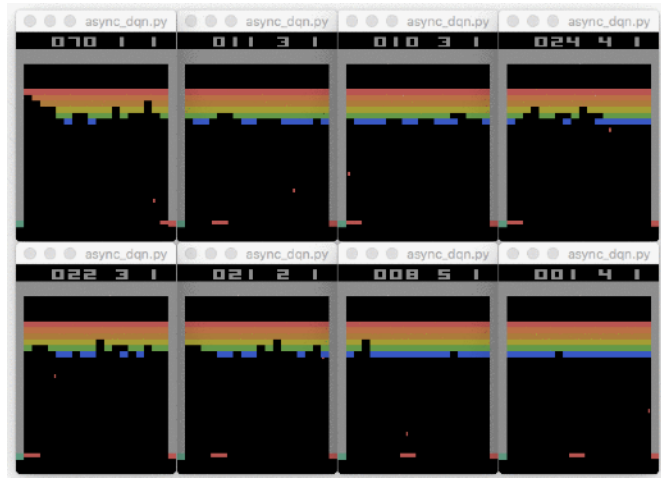


- Conduct **action** in any **state** of an **environment**.

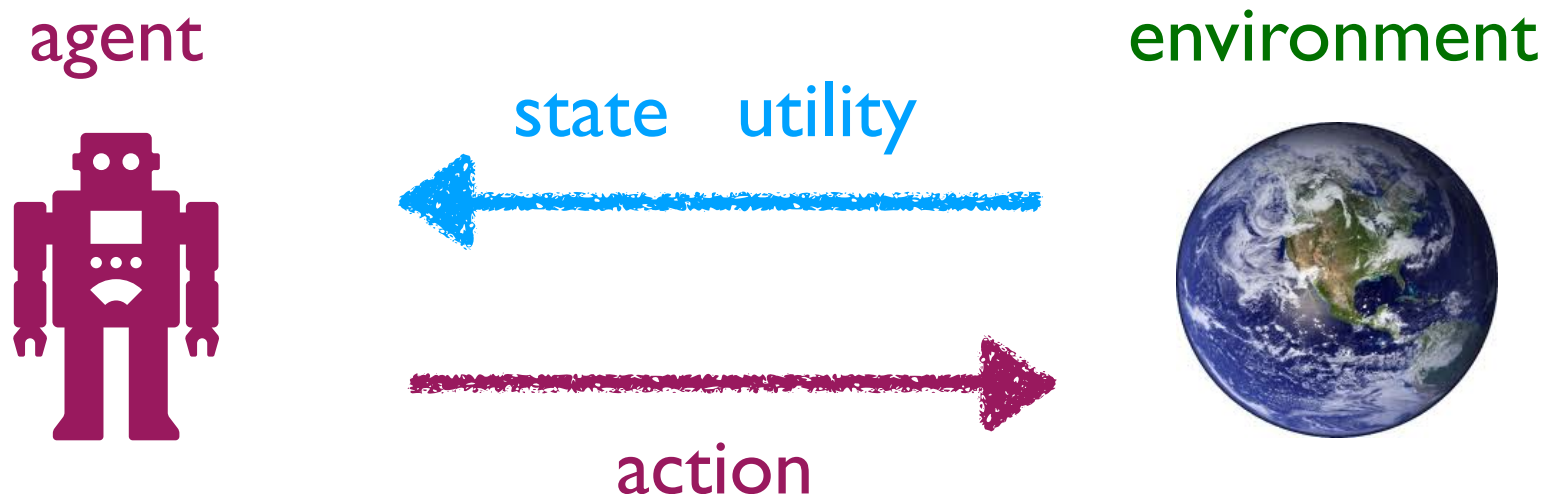


In most problems, the agent needs to do a sequence of actions w.r.t. a sequence of states.

# Decision Making

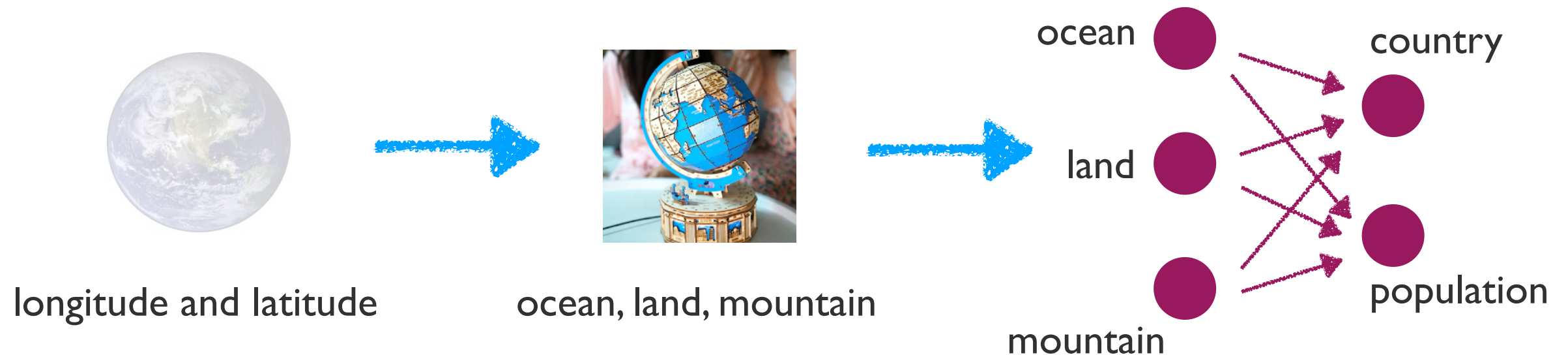


- Conduct **action** in any **state** of an **environment**.



In most problems, the agent needs to do a sequence of actions w.r.t. a sequence of states.

# Knowledge in AI Systems



- Turn primitive external states into meaningful internal states.
- Reason about most useful states for decision making.
- Capture internal relationships among factors of decision making.

These reasoning rules are called knowledges in an AI system.

# Turing Test

*“The new form of game can be described in terms of a game which we call the ‘imitation game’”.*

*“Instead of trying to produce a programme to simulate the adult mind, why not rather try to produce one which simulates the child’s?”*

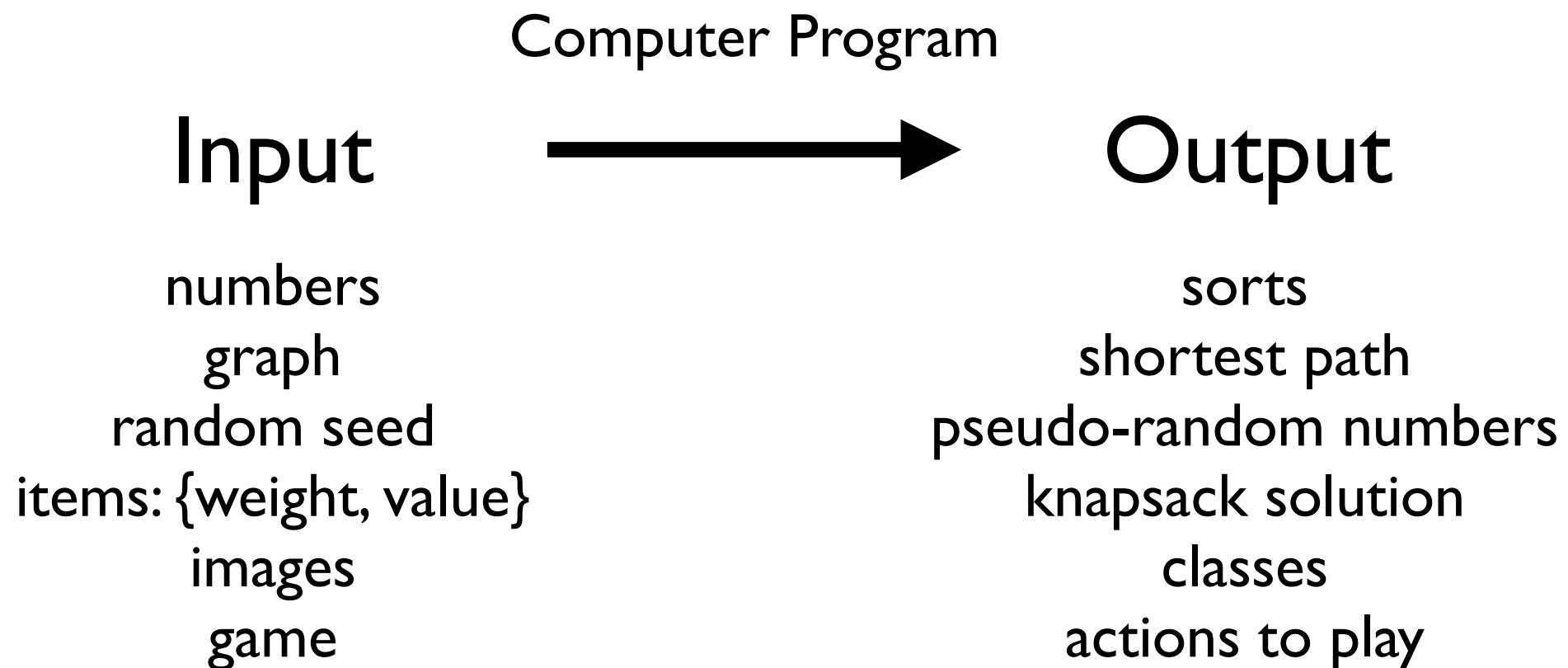
— Alan Turing, “Computing Machinery and Intelligence”, 1950.



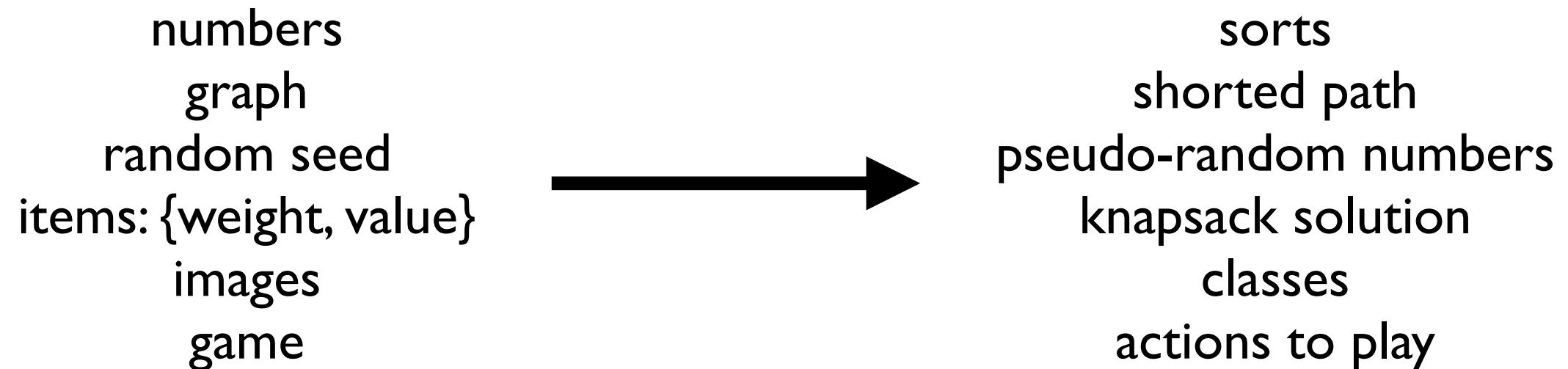
# Machine Learning: A Prelude

# Machine Learning

In machine learning, we want to obtain these programs (functions) by learning from experience instead of programming by hand.



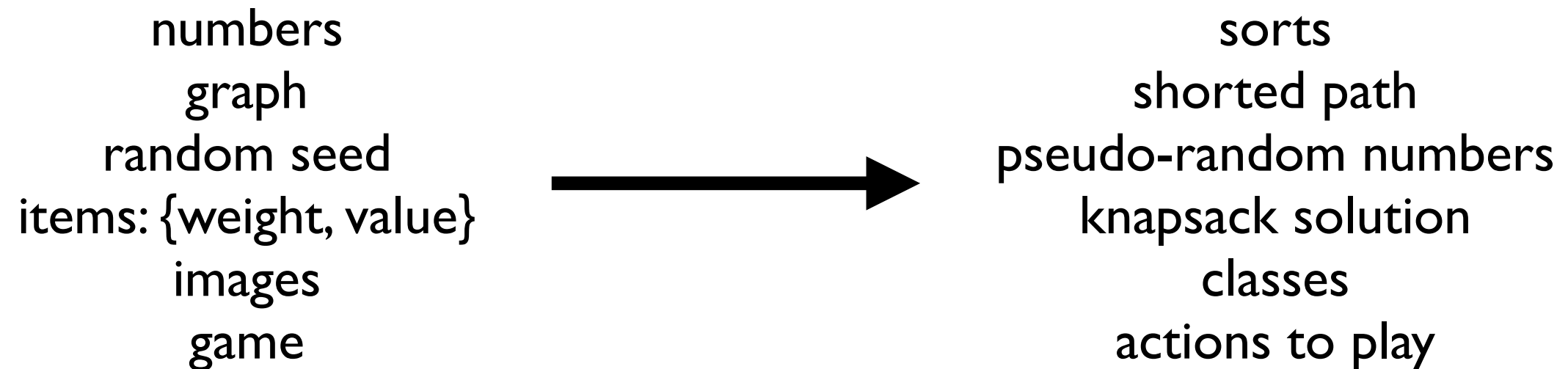
# Machine Learning



- Do all computer programs need to be obtained by machine learning?

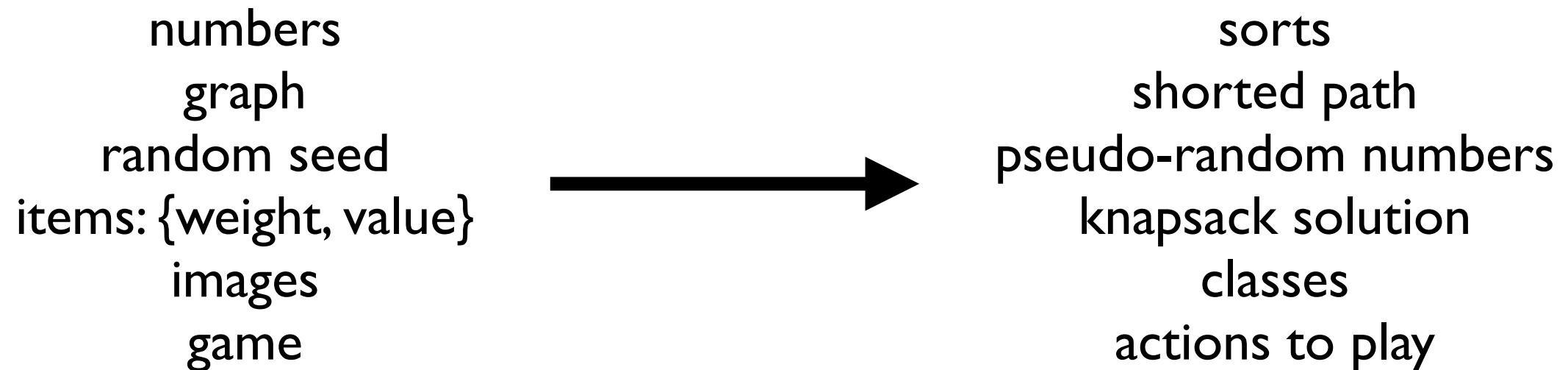


# Machine Learning



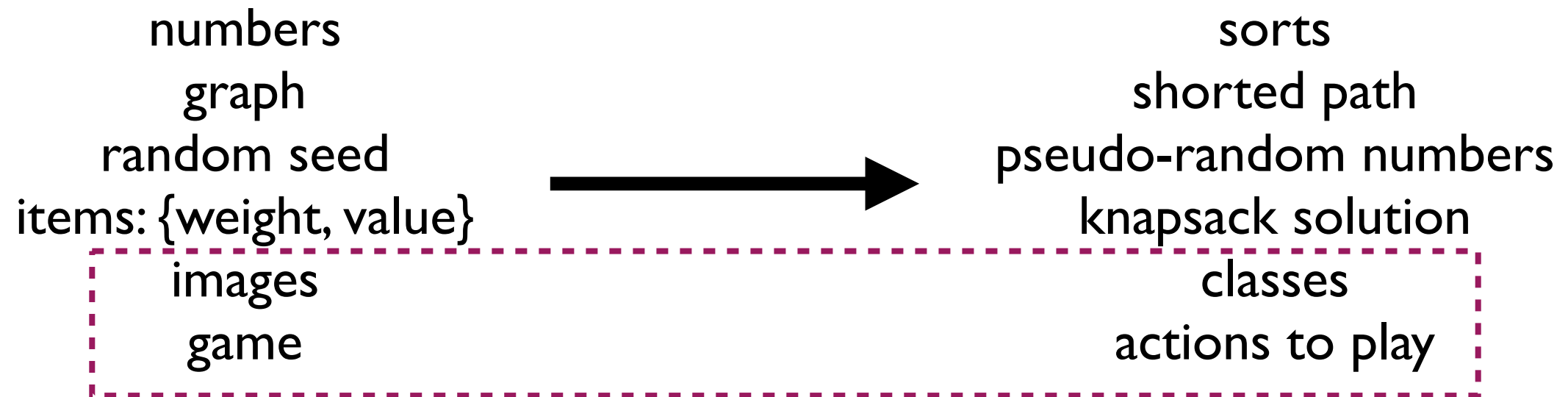
- Do all computer programs need to be obtained by machine learning?
- When learning is hard but programming-by-hand is easy: NO!

# Machine Learning



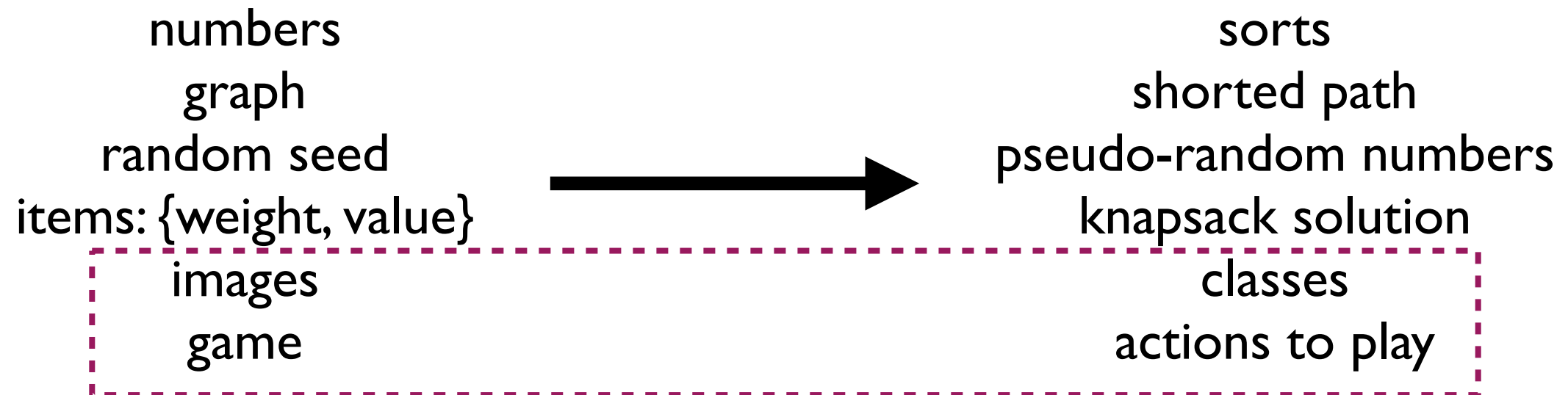
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- When programming-by-hand is hard but learning is easy: YES!

# Machine Learning



- Do all computer programs need to be obtained by machine learning?
- When learning is hard but programming-by-hand is easy: NO!
- When programming-by-hand is hard but learning is easy: YES!

# Machine Learning



- Do all computer programs need to be obtained by machine learning?
- When learning is hard but programming-by-hand is easy: NO!
- When programming-by-hand is hard but learning is easy: YES!

When will learning be possible?  
When learning is possible, how to learn efficiently?  
Similar to fundamental problems of computation.  
What's the difference?

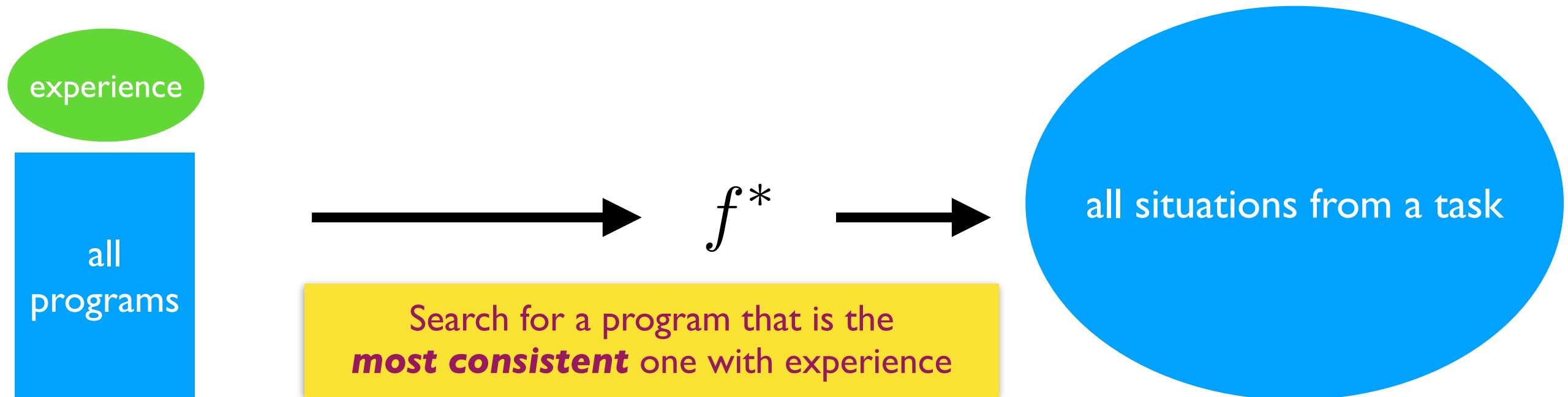
# Basic Mechanism



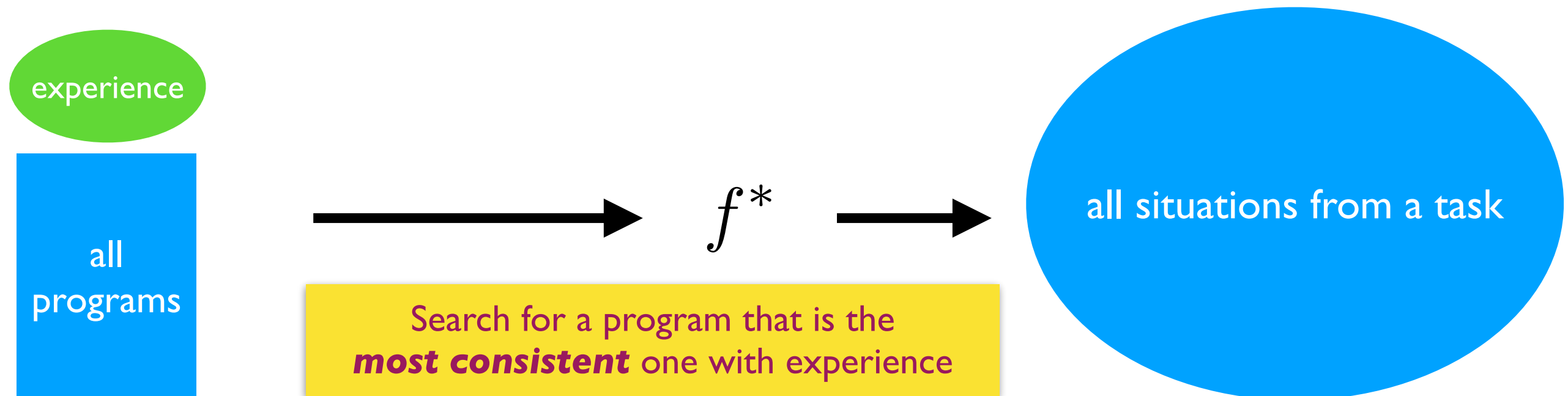
# Basic Mechanism



# Basic Mechanism



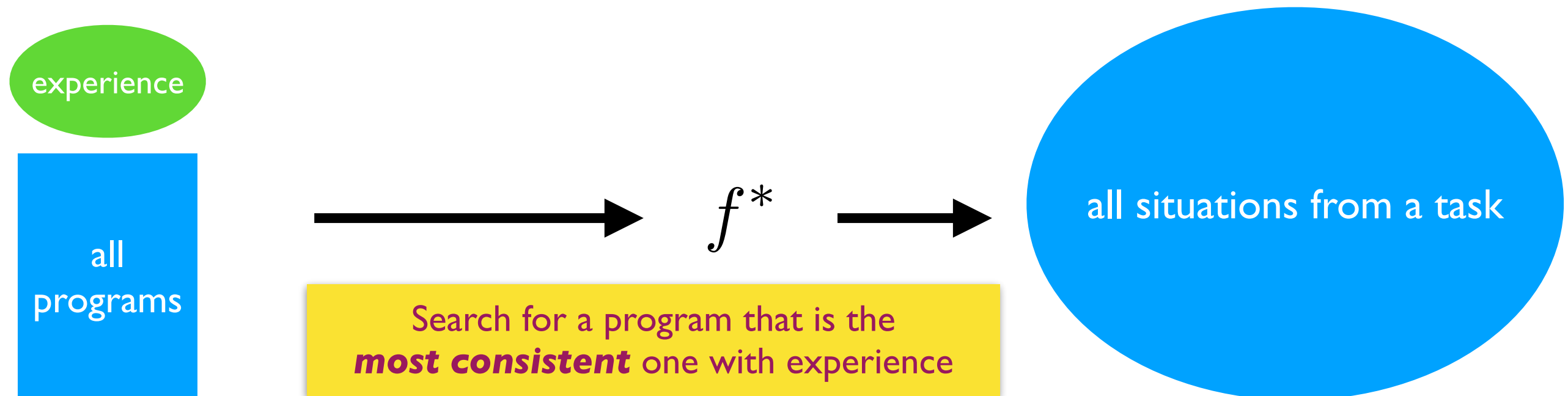
# Basic Mechanism



- When will learning be possible and easy?
  - When limited experience can represent all situations.
  - When searching for the best program can be done efficiently.

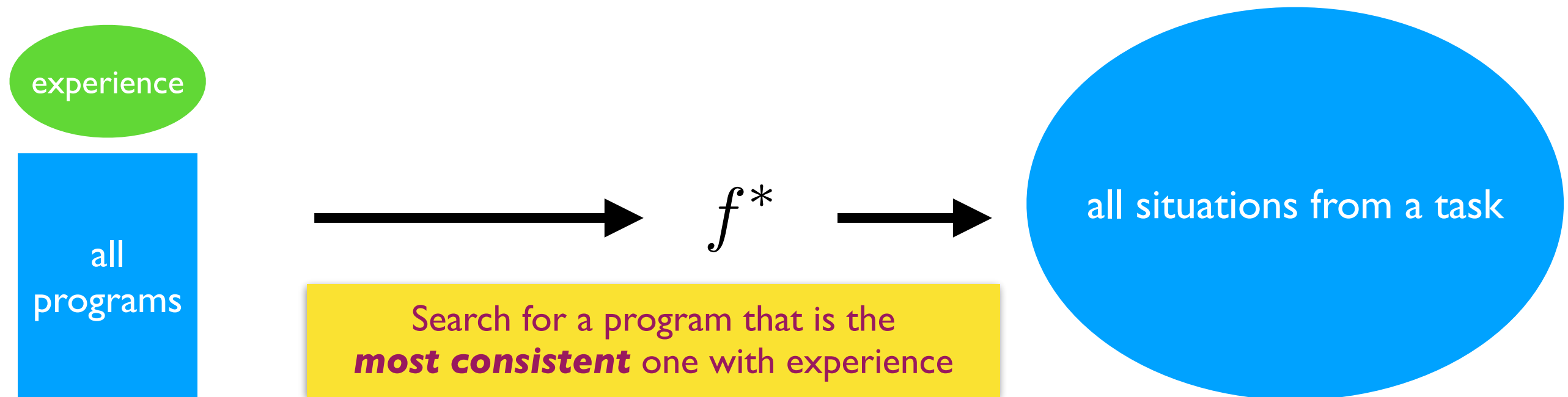


# Machine Learning is a **Statistical** Problem



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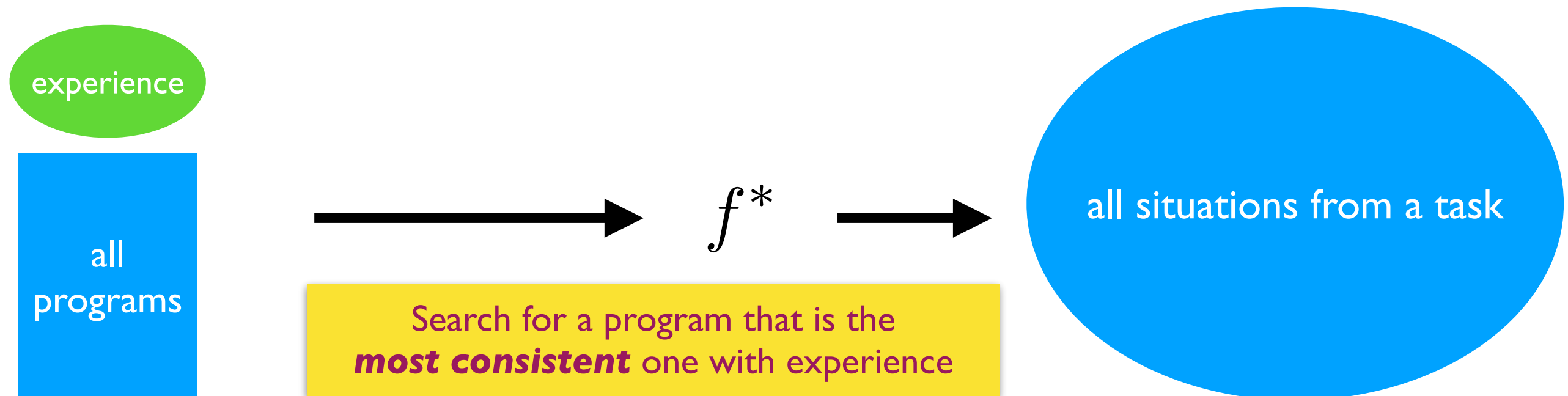
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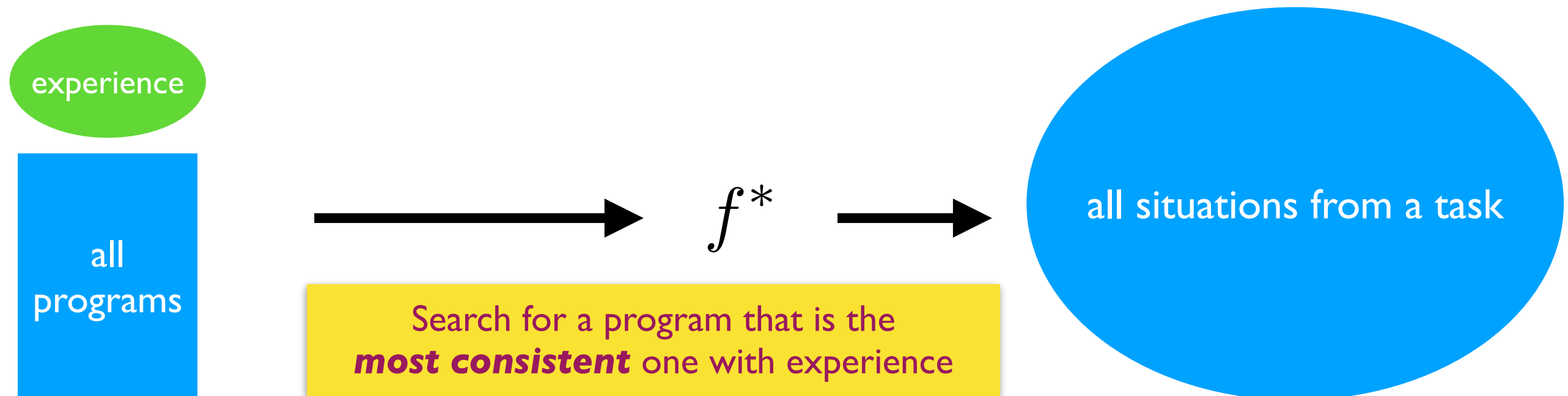
Obtain general rule from limited experiences: **statistics**.  
Basic principle: the law of large numbers.

# Machine Learning is a **Computation** Problem



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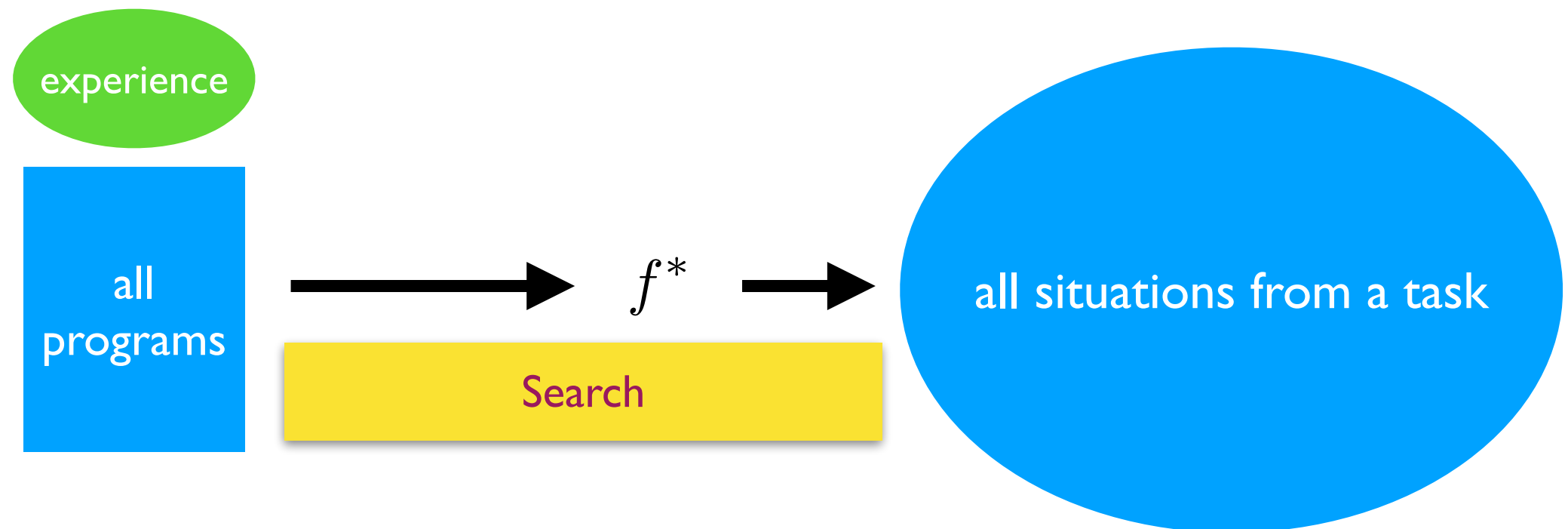
# Machine Learning is a **Computation** Problem



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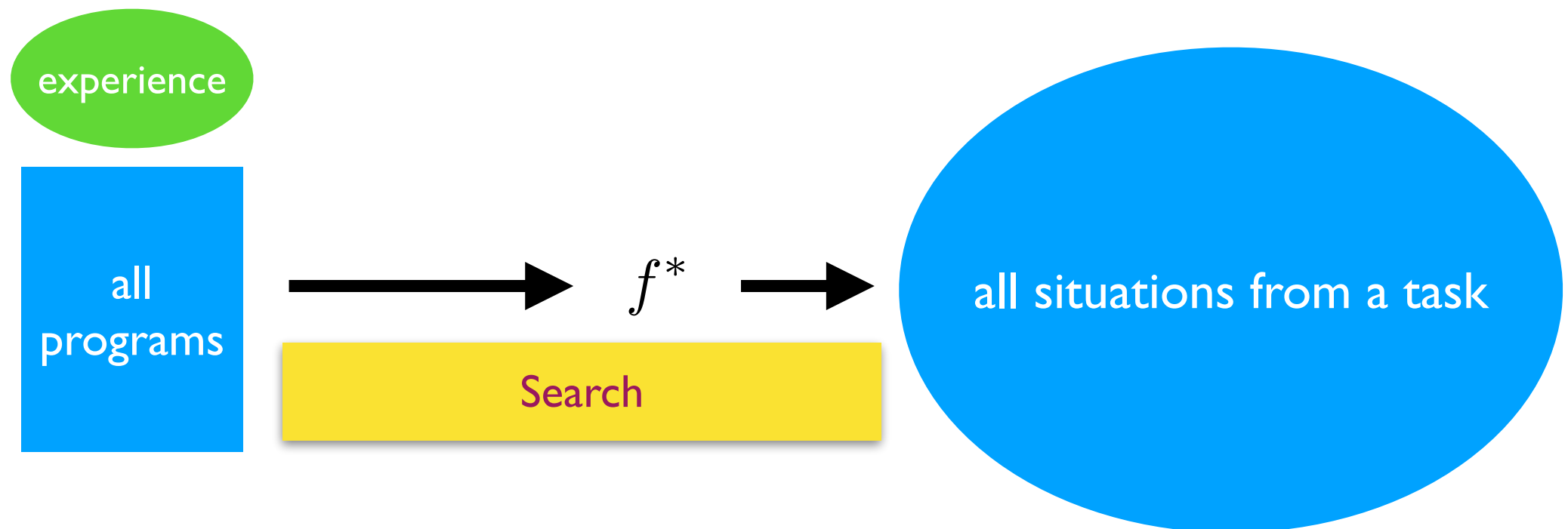
Searching from the optimal solution: **Optimization**.  
Optimization is a computation problem.

# Inductive Bias in Machine Learning



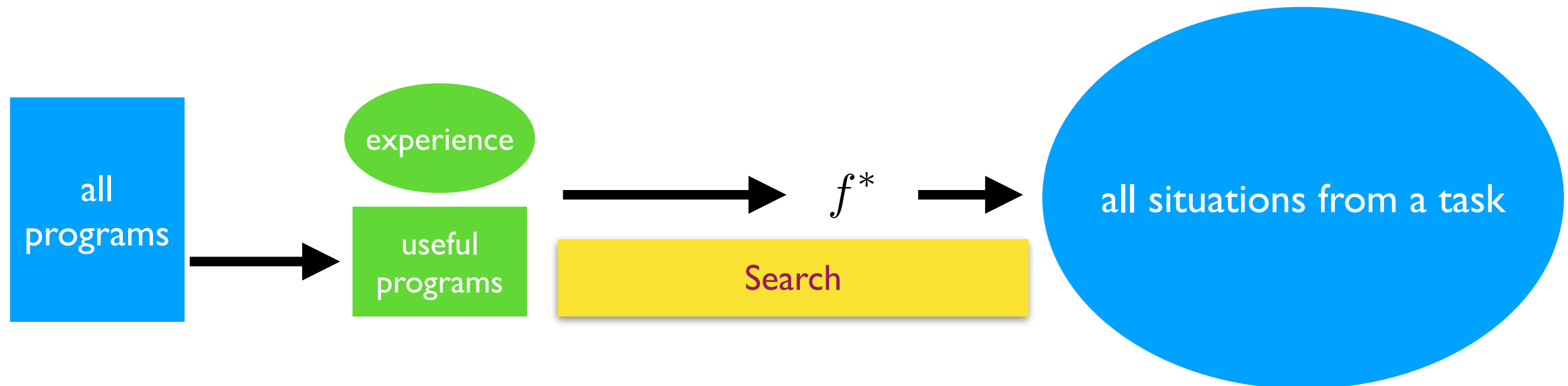
- Learning is both statistical and computational.
- We care about both **statistical and computational** complexity.

# Inductive Bias in Machine Learning



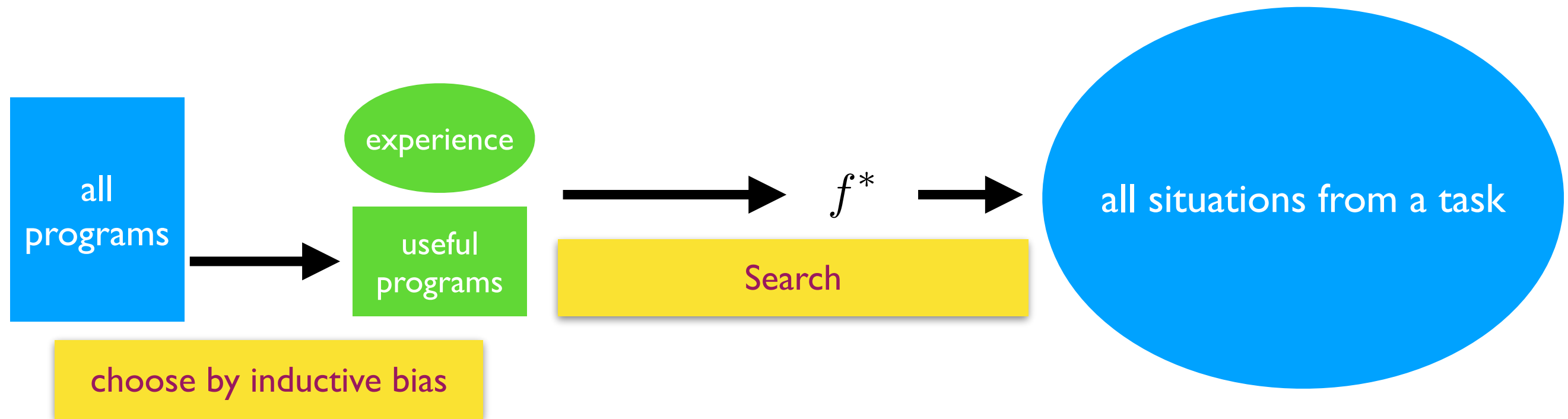
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- Humans can learn from very few experiences and very fast. Why?

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# Inductive Bias in Machine Learning



- Learning is both statistical and computational.
- We care about both **statistical** and **computational** complexity.
- Humans can learn from very few experiences and very fast. Why?

Not all programs are born equal.  
Humans choose the right **inductive bias** to reduce the program set.



# Machine Learning: Theory, Algorithm, and Application

- Machine Learning Theory:

- Understanding the foundations: when will learning be easy?

theoretical computer science, statistics, game theory, information theory...

- Machine Learning Algorithm:

- Design models and algorithms for *general* machine learning problems.

- Machine Learning Application:

- Design inductive bias for different applications, e.g. CV, NLP.

# Machine Learning Scenarios

- Supervised learning
  - Classification
  - Regression
  - Ranking
- Unsupervised learning
  - Clustering, density estimation
  - Generative modeling
- Semi-supervised learning
- Reinforcement learning
- ...

# Machine Learning Scenarios

- Supervised learning
  - Classification
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- Unsupervised learning
  - Clustering, density estimation
  - Generative modeling
- Semi-supervised learning

← learn from pre-given data

- Reinforcement learning
- ...

← learn from self-generated data

# Machine Learning Methods

- Symbolic Learning
- Frequentist statistical learning
  - Support vector machine, kernel method
  - Decision tree, random forest, boosting
- Bayesian statistical learning
  - Graphical models
  - Variational inference and approximate sampling
- Neural networks
  - Deep learning

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- Neural networks
  - Deep learning

No absolute boundaries!

# Take-Home Messages

- Machine learning is to search for a program that is the most consistent one with experience: statistics + computation.
- There are two fundamental questions to answer:
  - When will learning be easy?
  - When learning is easy, how to learn efficiently?
- Learning is easy with proper:

**Inductive Bias + Experience + Optimization**

# Recommended Textbooks



# Schedules

- We will have six lectures on machine learning
  - Two lectures on statistical learning
    - Nearest neighbor vs. linear classification
    - SVM & voting classifiers
  - Two lectures on neural networks and deep learning
    - logistic regression, perceptron, multi-layer neural networks
    - Deep learning
  - One lecture on generative deep models: autoencoder & GAN
  - One lecture on (more advanced) reinforcement learning



Thanks for your attention!  
Discussions?